

1 **Rapid earthquake-tsunami modeling: The multi-event,**
2 **multi-segment complexity of the 2024 M_W 7.5 Noto**
3 **Peninsula Earthquake governs tsunami generation**

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9 **Key Points:**

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- 11 The earthquake ruptures bilaterally, including six subevents and a re-nucleation
episode at its hypocenter 20 seconds after its initiation.
- 12 Our complex subevent model aligns with known fault system geometries and is
critical in explaining the observed tsunami.
- 13 Our simulation matches tsunami wave amplitude, timing, and polarity of the lead-
ing wave, which are crucial for tsunami early warning.

14

15

Abstract

The January 1st, 2024, moment magnitude (M_W) 7.5 Noto Peninsula earthquake ruptured in complex ways, challenging timely analysis of the tsunami generation. We present rapid and accurate tsunami models informed by a 6-subevent centroid moment tensor (CMT) model that we obtain by inverting teleseismic and strong motion data and validation against geodetic observations. We identify two distinct bilateral rupture episodes, including six subevents and a re-nucleation episode at its hypocenter 20 seconds after its initiation, likely aided by fault weakening. We construct a complex uplift model that aligns with known fault system geometries and is critical in modeling the observed tsunami. Our tsunami simulation can explain wave amplitude, timing, and polarity of the leading wave, which are crucial for tsunami early warning. Analyzing a 2000 multi-CMT solution ensemble and comparing to alternative rapid source models, we highlight the importance of incorporating complex source effects for realistic tsunami simulations.

Plain Language Summary

The 2024 moment magnitude 7.5 New Year's Day Noto Peninsula earthquake ruptured a complex, partially offshore fault system and generated a tsunami in the Sea of Japan. We use seismic data to show that the earthquake can be characterized by six distinct subevents, with an initial predominantly onshore rupture propagation towards the southwest and a 20-second delayed second rupture onset towards the northeast, mostly offshore. We use the information we gain from these subevents, such as location and faulting mechanism, to infer the seafloor movement, which informs tsunami simulations. The reconstruction of the earthquake rupture process is not unique. This allows us to explore the influence of source uncertainties on the modeled tsunami, highlighting the importance of complex source effects for tsunami generation. We find that our preferred model matches tsunami onset times, first-motion polarities, and initial wave amplitudes, crucial aspects for tsunami early warning.

1 Introduction

The January 1st, 2024 M_W 7.5 Noto Peninsula (Noto-Hanto) earthquake ruptured an active submarine fault system (Fig. 1, MLIT (2014); Sato et al. (2020)) causing strong ground shaking and a large tsunami within the Sea of Japan. Early analysis points to an unusually complex rupture process, with rapidly estimated slip distributions differing considerably (Fujii & Satake, 2024; Masuda et al., 2024; Okuwaki et al., 2024; U.S. Geological Survey, 2024).

Rapid finite-fault models based on teleseismic data were available within hours after the event (The Headquarters for Earthquake Research Promotion, 2024; U.S. Geological Survey, 2024). The United States Geological Survey (USGS) released a first version obtained solely from the teleseismic data (hereafter model “USGS-T”, Fig. S1). Later, the USGS released an updated model using both the teleseismic and Global Navigation Satellite System (GNSS) data (hereafter model “USGS-T+G”, Sec. 2.3). This model differs starkly from the earlier version. Specifically, the updated USGS-T+G model does not have a significant offshore slip.

Another finite-fault model is obtained using 53 GNSS stations across the Noto Peninsula, placing the majority of slip onshore or near the northern shoreline of Noto Peninsula (Fujii & Satake, 2024). In contrast to the USGS-T+G model, a finite-fault model from tsunami waveforms recorded around the Sea of Japan places most of the slip offshore (Fujii & Satake, 2024). Additionally, Masuda et al. (2024) investigated landslide contributions to local tsunami generation, but precise reconstruction is challenged by the limited regional bathymetry resolution. Source complexity is important for tsunami generation and propagation (Abrahams et al., 2023; Dettmer et al., 2016; Lotto et al., 2018;

65 Wirp et al., 2021). Thus, vastly different source models will have different implications
66 for understanding the observed tsunami generation and early warning.

67 This study aims to address the challenge of rapidly and robustly resolving earth-
68 quake rupture complexities and properly translating those complexities to inform accu-
69 rate tsunami simulations. We present rapid and accurate tsunami simulations construct-
70 ing complex seafloor displacements from a Bayesian 6-subevent centroid moment ten-
71 sor (CMT) model that we obtain using teleseismic and strong motion observations of the
72 Noto Peninsula earthquake. We unify seismic and tsunami observations in agreement
73 with geodetic data. While CMT solutions are rapidly available, they are rarely used in
74 routine tsunami modeling or early warning contexts (Gusman & Tanioka, 2014; Miyoshi
75 et al., 2015). To the best of our knowledge, this study is the first to use a multi-CMT
76 model to source a tsunami simulation. We demonstrate that our approach captures key
77 characteristics of the tsunami complexities better than other rapid finite-fault inversion
78 approaches and discuss the effects of source complexity and its uncertainties on tsunami
79 modeling based on an ensemble of 2000 multi-CMT solutions.

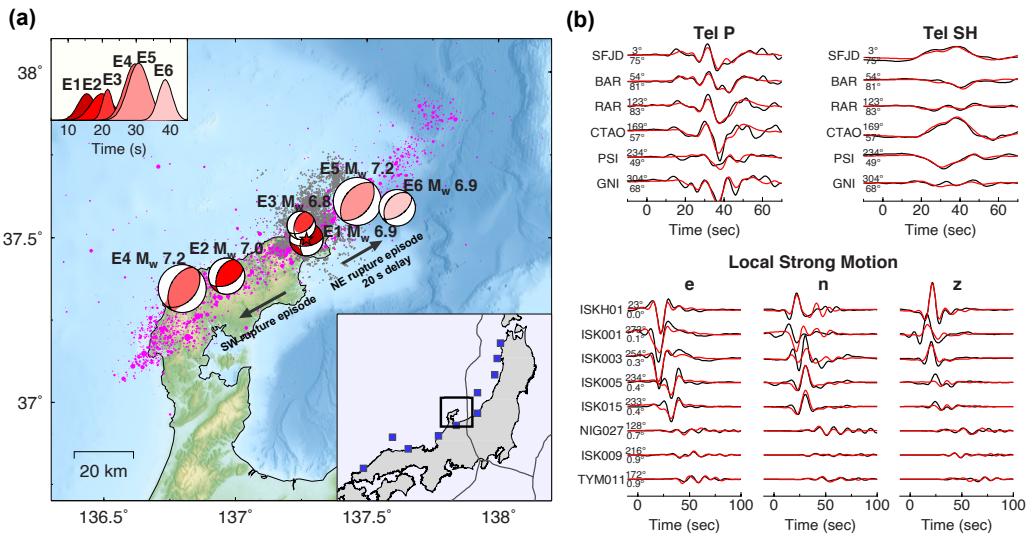


Figure 1. (a) Overview of the Noto Peninsula, Japan, study area. The red star indicates the JMA epicenter of the January 1, 2024, M_w 7.5 Noto Peninsula earthquake. The red focal mechanisms are the six subevents of the Bayesian multi-centroid moment tensor (CMT) inversion using teleseismic and regional strong motion data. The earthquake first initiates towards the southwest, indicated by subevents E1, E2, and E4. After a delay of 20 s, the rupture unfolds towards the northeast, as indicated by subevents E3, E5, and E6. The focal mechanisms are color-coded with respect to time, and the corresponding Gaussian source time durations are shown in the top left figure inset. The blue squares in the bottom right figure inset mark the position of tide gauges facing the Sea of Japan. (b) Comparison of selected observed teleseismic P, SH, and local strong ground motion recordings (black) with the corresponding synthetic seismic waveforms (red) of the preferred multi-CMT solution. The numbers leading the traces are the respective azimuth and distance. A complete comparison of all seismic waveforms is shown in Figs. S2-S7.

80 **2 Methods**81 **2.1 Seismic Multi-Centroid Moment Tensor Inversion**

82 We constrain the event's rupture propagation using a multiple CMT subevent in-
 83 version method (Tsai et al., 2005; Minson & Dreger, 2008; Jia et al., 2022, 2023). We
 84 iteratively increase the number of subevents to achieve a 65% waveform misfit reduction
 85 (Figs. S2-S6). The preferred model includes six subevents, E1 to E6, ordered by their
 86 centroid time (Fig. 1). Each subevent is characterized by 10 unknowns: centroid time,
 87 duration, longitude, latitude, depth, and the five independent components of the sym-
 88 metric and zero-traced moment tensor (Fig. S8, Table S1). We fix the longitude and lat-
 89 itude of the first subevent at the JMA (Japan Meteorological Agency) epicenter loca-
 90 tion while solving for its depth.

91 We use a Metropolis–Hastings Markov Chain Monte Carlo (MCMC) method to sam-
 92 ple the posterior probability density function in a Bayesian framework. This MCMC in-
 93 version first searches the centroid time, duration, longitude, latitude, and depth and then
 94 linearly solves for the independent moment tensor components (Jia et al., 2020, 2022,
 95 2023). As a prior constraining subevent locations, we use the horizontal spatial density
 96 of the first three days of aftershocks. We run the MCMC inversion for 15,000 iterations
 97 and consider the first 10,000 samples as the burn-in period. This burn-in process ensures
 98 a misfit convergence with fluctuations of less than 10% from the optimal model. The ran-
 99 domness of the initial model can significantly influence the MCMC convergence, with
 100 an unfortunate draw potentially trapping a chain in a local minimum (Olalotiti-Lawal
 101 & Datta-Gupta, 2018; Ray et al., 2013). Therefore, we remove the least-performing 75%
 102 of chains, retaining the best-fitting 48 of all 192 chains to reflect the primary posterior
 103 probability features. In total, we obtain an ensemble of 240,000 permissible multi-CMT
 104 solutions.

105 We choose the preferred multi-CMT model based on minimizing the seismic wave-
 106 form misfit. We use 93 teleseismic P and SH waveforms in an epicentral distance range
 107 of 30° to 90° obtained from the EarthScope Data Management Center (DMC; Albuquerque
 108 Seismological Laboratory/USGS, 2014). Additionally, we use 53 three-component regional
 109 KIK-net and K-net strong ground motion waveforms within an epicentral distance of 150 km
 110 from National Research Institute for Earth Science and Disaster Prevention (NIED; Okada
 111 et al., 2004). We remove the instrument response and detrend the data. We bandpass
 112 filter the teleseismic waveforms between 0.005–0.2 Hz and the strong ground motion data
 113 between 0.01–0.1 Hz. During the inversion of regional strong motion data, we adopt the
 114 JMA2001 1D velocity model (Ueno, 2002), and use a frequency-wavenumber method (L. Zhu
 115 & Rivera, 2002) to calculate the Green's functions. For the inversion of teleseismic waves,
 116 we calculate the Green's functions with a hybrid method that combines propagator ma-
 117 trix and ray theory (Kikuchi & Kanamori, 1982; Qian et al., 2017), and use a combina-
 118 tion of the JMA2001 model for the crust with an IASPEI91 model (Kennett & Engdahl,
 119 1991) describing the deeper earth.

120 **2.2 Mapping the Subevent Model to Seafloor Deformation**

121 We construct a six-fault-segment slip model based on the preferred subevent model
 122 (Table S2), assuming rectangular faults. Each fault segment is located at the respective
 123 subevent centroid location. We determine their dip, strike, and rake angles from the pre-
 124 ferred multi-CMT solution. E1-E5 are considered southeast dipping, and E6, located in
 125 the northeast of Noto Peninsula, dips towards the northwest. Each fault segment has
 126 an along-strike length of 25 km and extends from the surface with an along-dip depth
 127 twice its centroid depth.

128 Informed by the preferred multi-CMT model, we assume a uniform slip distribu-
 129 tion across each of the six fault segments, which we obtain from each respective subevent's

130 seismic moment and an assumed rigidity of 35 GPa, which resembles the mean rigidity
 131 of the shallowest 25 km as given by the JMA2001 velocity model (Ueno, 2002) and is sim-
 132 ilar to the assumed value in Fujii and Satake (2024); Masuda et al. (2024). We then use
 133 an analytic elastic dislocation model (Okada, 1985, 1992) to obtain the corresponding
 134 surface displacements.

135 To evaluate the uncertainties in surface deformation and its impact on tsunami gen-
 136 eration, we repeat this analysis for 2000 randomly selected realizations out of the 240,000
 137 MCMC ensemble solutions (Fig. S9, Table S3). We use the sum of the absolute offshore
 138 vertical displacement due to the 2000 multi-CMT solutions as a metric to identify two
 139 endmember multi-CMT solutions, the minimum and maximum uplift CMT solutions,
 140 which yield the least and the most amount of offshore vertical displacements (Fig. S10),
 141 respectively.

142 2.3 USGS Finite-Fault Models

143 We compare the surface deformation and resulting tsunami of the multi-CMT so-
 144 lution to two USGS finite-fault models (U.S. Geological Survey, 2024). The USGS pub-
 145 lishes finite-fault models for $M \geq 7$ earthquakes within a few hours (Thompson et al., 2019).
 146 These models are crucial for rapid response management and urgent seismic and tsunami
 147 hazard assessment. For the 2024 Noto earthquake, the USGS released an initial version
 148 of a finite-fault model (USGS-T), which is obtained from broadband teleseismic P, SH,
 149 and long-period surface waves after 2.5 hours (Ji et al., 2002). On 10 January 2024, the
 150 USGS released an updated finite-fault model (USGS-T+G), which incorporates addi-
 151 tional regional GNSS observations in the inversion (Goldberg et al., 2022). We compute
 152 surface deformation due to both USGS finite fault models, following the same Okada ap-
 153 proaches for our multi-CMT solution (Sec. 2.2). These models are then used to simu-
 154 late tsunamis and compared to our preferred model.

155 2.4 Tsunami Simulations

156 We use GeoClaw and the vertical offshore surface deformation as instantaneous sources
 157 for tsunami simulations. GeoClaw is part of the open-source software package ClawPack
 158 (LeVeque et al., 2011; Berger et al., 2011; Mandli et al., 2016), which solves the 2D depth-
 159 averaged shallow water equations and has been validated against community benchmark
 160 problems and real observations (LeVeque & George, 2008; González et al., 2011; Arcos
 161 & LeVeque, 2015). The algorithm has been successfully applied to model the 2004 Suma-
 162 tra tsunami (Ulrich et al., 2022) and the 2017 Tehuantepec tsunami in Mexico (Melgar
 163 & Ruiz-Angulo, 2018).

164 We utilize space-time adaptive mesh refinement with a maximum level of 3 for all
 165 tsunami simulations. This results in a maximum spatial resolution of 493 m. We use grid-
 166 ded bathymetry data (GEBCO Compilation Group, 2023) with a resolution of 15 arc
 167 seconds (450 m). We note that without rapidly available high-resolution bathymetry and
 168 due to interpolation differences in station locations obtained from the Flanders Marine
 169 Institute (VLIZ), Intergovernmental Oceanographic Commission (IOC) (2024) and Geospatial
 170 Information Authority of Japan (GSI) (2024), we move the location of all tidal stations
 171 besides stations Tobishima and Oga to the nearest “wet” cell (Wang et al., 2020), which
 172 has minimal impact on our tsunami results. We define the sea surface height anomaly
 173 (ssha) as the deviation from the ocean surface at rest. We simulate all tsunami scenar-
 174 ios for three hours, which requires ~ 7.5 h per simulation on a laptop (MacBook Air with
 175 Apple M2 processor).

176 We validate our simulated tsunami waveforms with sea level observations obtained
 177 from the IOC and the GSI. First, we use the LOWESS algorithm (Locally Weighted Scat-
 178 terplot Smoothing; Cleveland, 1979; Romano et al., 2021) to remove first-order tidal trends.

179 Next, we trim the data to three hours after the mainshock origin time (2024-01-01 7:10:22.5
 180 UTC; provided by the JMA) before applying a 300 s lowpass filter. To quantify the sim-
 181 ilarity of the simulated and observed first-arriving wave packet at all tide gauges, we cal-
 182 culate the cross-correlation coefficient for a 20 min time window, starting 5 min before
 183 the arrival of the peak of the initial tsunami crest (Table S4).

184 3 Results

185 3.1 Multi-event, Multi-segment Rupture of the 2024 M_W 7.5 Noto Earth- 186 quake

187 Our subevent model reveals two distinct rupture episodes (Fig. 1). Initially, rupt-
 188 ture propagates towards the southwest (subevents E1, E2, and E4), lasting for about 30
 189 seconds. Following a delay of 20 s, while the southwest rupture is ongoing, the rupture
 190 re-nucleates around the hypocenter (E3) and propagates bilaterally towards the north-
 191 east direction (E5 and E6) for 15 seconds.

192 The geometry of our preferred six-fault-segment slip model aligns with regional mapped
 193 fault traces (Fig. S10; Fujii & Satake, 2024; MLIT, 2014) and spatially coincides with
 194 the first 30-hour aftershock sequence (Movie S1). The hypocentral subevents E1 and E3
 195 are collocated with four year swarm activity preceding the Noto earthquake (Hubbard
 196 & Bradley, 2024; Nishimura et al., 2023; Yoshida et al., 2023).

197 These six subevents share similar reverse-faulting focal mechanisms, albeit vary-
 198 ing significantly in size and duration. The nucleation and re-nucleation subevents, E1
 199 and E3, have the smallest moment magnitudes (both M_W 6.9). The two largest subevents,
 200 E4 and E5, each with M_W 7.2, are located near the two endpoints of rupture. Subevents
 201 E2, E4, and E5 each have a source duration of \sim 13 s, while the duration for the other
 202 three subevents is shorter and ranges between \sim 6-11 s.

203 Robust estimates of event depth and fault geometry are critical for simulating the
 204 surface deformation and associated tsunami. Using the ensemble of 240,000 multi-CMT
 205 solutions, we analyze source parameter uncertainties. We find that the subevent depths
 206 are well-constrained (\leq 10 km) for all subevents (Fig. S8), with an average standard de-
 207 viation of 1.17 km. All subevent focal mechanisms, except that of E3, also exhibit low
 208 uncertainties in strike, dip and rake, with average standard deviations of 15.9° , 4.9° , and
 209 21.3° , respectively (Fig. S9 and Table S3). The geometry of the re-nucleation subevent
 210 E3 has distinctly larger uncertainties, with 88.9° , 14.7° , and 101.1° , in strike, dip, and
 211 rake, which likely arise from its concurrence with ongoing southwest rupture, challeng-
 212 ing resolution. However, subevent E3 is necessary to explain the closest strong motion
 213 waves (Fig. S7).

214 3.2 Complex Onshore and Offshore Surface Deformation

215 The synthetic surface displacements due to the complex rupture of the Noto earth-
 216 quake show a peak vertical offshore uplift of 3.91 m. Our synthetic surface deformation
 217 match the regional GNSS observations, showing a broad uplift across the northern Noto
 218 Peninsula and subtle subsidence in the far-field (Fig. 2a). Subevents E1-E4 result in a
 219 combination of onshore and offshore surface deformation, while the uplift generated by
 220 subevents E5 and E6 is located entirely offshore (Fig. 2). The respective northeast rup-
 221 ture episode releases 40% of the seismic moment, translating into up to 5.27 m of off-
 222 shore fault slip.

223 Our model predicts less vertical motion than the one recorded at station J576. How-
 224 ever, both the USGS-T+G model and the finite-fault model from Fujii and Satake (2024)
 225 cannot fully capture the amount of observed subsidence at this site either (e.g., Fig. S1),

226 suggesting it may be affected by local processes such as landslides or liquefaction (Gomez,
227 2024; Kataoka et al., 2024).

228 Our predicted subevent surface displacement produces substantial vertical motion
229 offshore compared to the limited amount of uplift suggested by USGS-T+G and USGS-
230 T models (Fig. 2). The latter (Fig. S1) predicts an offshore vertical uplift up to 1.38 m
231 (Fig. 2b), while the USGS-T+G model (Fig. 2c) predicts a negligible amount of uplift
232 in the northeast of Noto Peninsula (Fig. S1). These differences directly affect the tsunami
233 simulations (Sec. 3.3).

234 We evaluate the effects of source parameter uncertainties on predicted surface dis-
235 placement and the associated tsunami simulations. We examine the surface deformations
236 caused by 2000 permissible multi-CMT solutions. The peak offshore uplift varies con-
237 siderably and has a standard deviation of 1.43 m (Fig. 4a). The minimum uplift CMT
238 model locates the subevents E1-E4 further landwards and produces a significantly re-
239duced offshore uplift of up to 3.02 m (Fig. S10b). The maximum uplift model locates
240 subevents E1-E4 mostly offshore, leading to a large offshore uplift of up to 4.26 m (Fig. S10c).

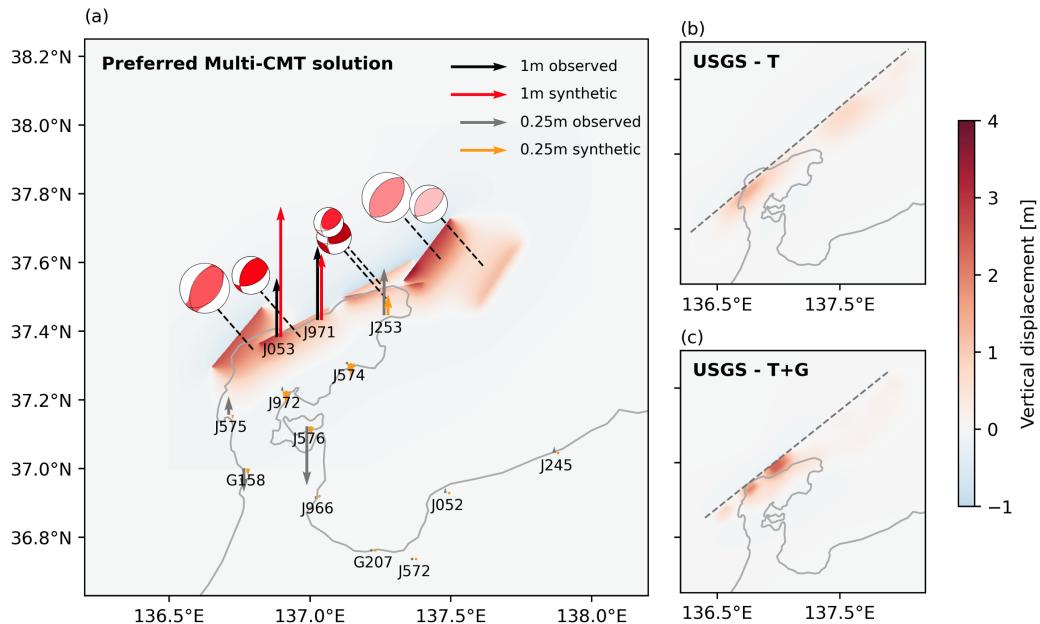


Figure 2. Synthetic vertical displacement constructed from (a) the preferred multi-CMT model, (b) the USGS-T, and (c) the USGS-T+G finite-fault models using an Okada approach. The color-coded vertical arrows show the observed and synthetic vertical displacement at GNSS sites, respectively. The six subevents of the preferred multi-CMT solution are indicated by their moment-tensor solutions. Grey lines in panels (b), (c) represent the fault trace of the respective USGS finite-fault model.

241 3.3 Complex Tsunami in the Sea of Japan

242 Our tsunami simulation shows complex coastal wave behavior (Movie S2), includ-
243 ing wave crests bending parallel to the coastline due to refraction at the shoaling bathymetry
244 (Fig. 3a). Our simulated tsunami waves capture the timing, initial polarity, and ampli-
245 tude of the first-arriving crest at all nine tide gauges shown in Fig. 3, and the overall shape
246 of the observed tsunami waveforms at most of them. Specifically, the timing, crucial for
247 tsunami early warning, is captured with high accuracy (within 1 to 3.5 minutes depend-

248 ing on station distance, Fig. 3b). We achieve overall high cross-correlation coefficients
 249 (Sec. 2.4) between the synthetics and observations during the first tsunami wave packet
 250 (Fig. 3b). However, it is challenging to fully capture the waveform complexity at the tide
 251 gauge Toyama (Fig. S11).

252 During the three hours of tsunami propagation modeled, our simulated amplitudes
 253 agreed with observations within six centimeters at Kashiwazaki, Tajiri, Oga, Saigo, and
 254 Okushiri stations. At Sado, Tobishima, and Fukaura stations, the fit of early waves is
 255 equally good but the model underestimates the amplitudes of later, trailing signals (Fig. 4b).
 256 The maximum tsunami wave amplitude distribution from our preferred simulation (Fig. 4a)
 257 indicates a large maximum tsunami height of up to 2.71 m in the source region. Our sim-
 258 ulation reveals long-lasting tsunami reverberations around the Noto Peninsula, appear-
 259 ing after 1 hour and 12 minutes (Fig. 3a, Movie S2). Such reverberations may be caused
 260 by trapped waves, causing energetic edge waves and/or shelf resonance, as has been ob-
 261 served during the tsunami caused by the M_W 8.2 Tehuantepec, Mexico, earthquake (Melgar
 262 & Ruiz-Angulo, 2018).

263 The tsunami simulation sourced by the minimum-uplift endmember of our source
 264 model ensemble underestimates tsunami amplitudes (peak 2.38 m; Fig. 4b, d, Fig. S12,
 265 Table S5). In distinction, the tsunami corresponding to the maximum uplift source yields
 266 a 24% larger peak tsunami height of up to 3.36 m compared to our preferred tsunami
 267 model (Fig. 4b e). Both rapidly available USGS source models generate localized tsunami
 268 (Fig. 4b, f, g), but neither can explain the observed tsunami amplitudes and timing (Fig. S13).

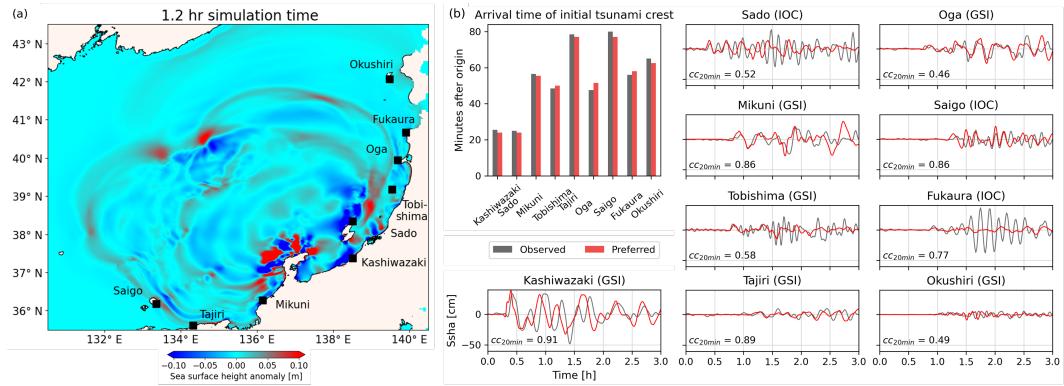


Figure 3. (a) Snapshot of tsunami propagation after 1 hour and 12 minutes simulated time, with strong tsunami reverberations surrounding the Noto Peninsula. At this point in time, towards the northeast, the tsunami reached the tide gauges Oga and Tobishima. In the southwest, the tsunami front arrives at tide gauges Saigo and Tajiri. (b) Comparison of observed and simulated tsunami arrival times and comparison of tsunami waveforms at nine tide gauges. Stations are ordered by their geodetic distance from subevent E1 (Fig. 1).

269 4 Discussion

270 An active seismic swarm preceded the M_W 7.5 Noto earthquake (Nishimura et al.,
 271 2023), recorded by a dense regional seismic network including events down to magnitude
 272 -3 (Hubbard & Bradley, 2024; Japan Meteorological Agency, 2024). Dominated by earth-
 273 quakes at depths of 14-16 km this swarm led to over 70 mm of surface uplift (Nishimura
 274 et al., 2023). Since November 2020, the swarm's activity has fluctuated, including a pe-
 275 riod of quiescence followed by a M_W 6.2 earthquake in May 2023, the largest event prior
 276 to the 2024 Noto earthquake (Kato & Nishimura, 2024). During the two weeks leading

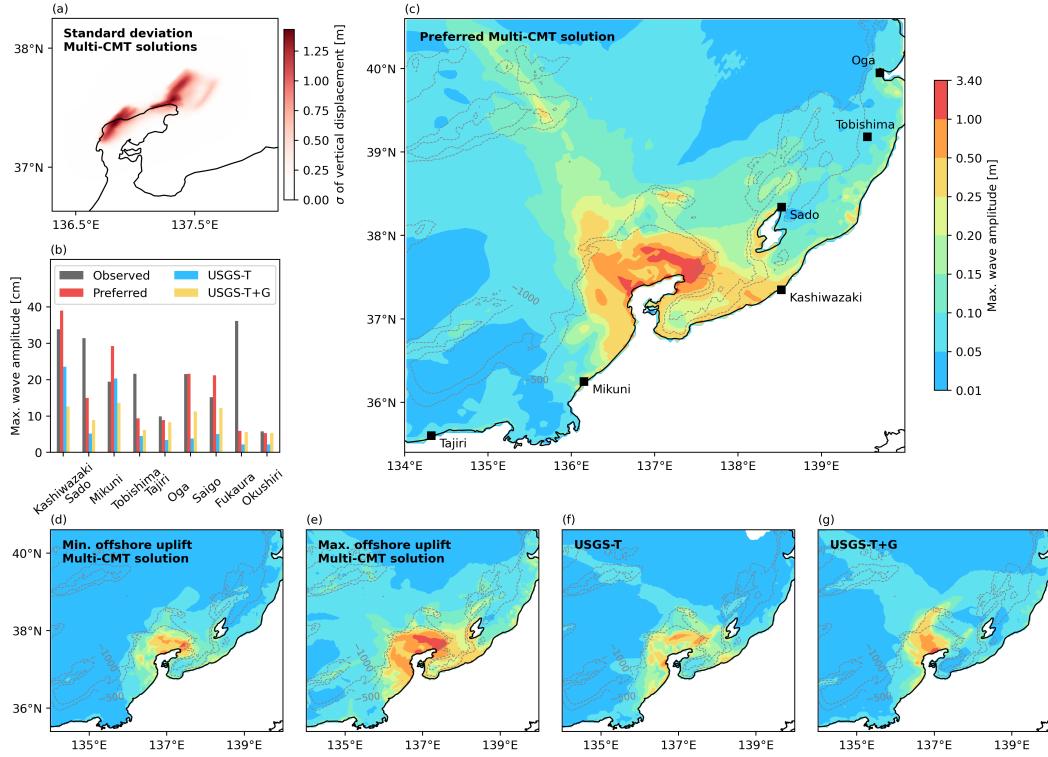


Figure 4. (a) Standard deviation of the vertical displacements based on an ensemble of 2000 multi-CMT solutions. (b) Histogram of the observed and simulated maximum wave amplitudes over a three-hour time window after the earthquake's origin time at the tide gauge locations shown in Fig. 3a. (c) Tsunami maximum wave amplitude distribution sourced by the preferred multi-CMT solution. (d), (e) Tsunami maximum wave amplitude distributions based on the minimum and maximum uplift multi-CMT solutions, respectively. (f), (g) Tsunami maximum wave amplitude distributions modeled using the USGS-T and USGS-T+G source models, respectively.

up to the main shock, a foreshock sequence developed, localizing within a 1 km radius of what would form the Noto earthquake's hypocenter within one hour before its origin time (Kato & Nishimura, 2024).

The spatial and temporal correlation between the swarm activity and the Noto earthquake may suggest a causal relationship with the event's complexity (Okuwaki et al., 2024; Yoshida et al., 2023). Our results indicate that the hypocentral region slipped more than once during the same earthquake, an effect that has been observed in laboratory experiments (Nielsen et al., 2010) and during other large earthquakes (Lee et al., 2006; Wald et al., 1990), including the 2011 Tohoku-Oki event (Lee et al., 2011; Yao et al., 2011) and that has been explained by weakened faults in theoretical and numerical analysis (Gabriel et al., 2012; Nielsen & Madariaga, 2003).

Earthquake swarms have been linked to aseismic slip or fluid migration (Lohman & McGuire, 2007; Ross et al., 2020). Related cyclic changes in pressure, permeability and fluid migration have been observed in a wide range of fault settings (e.g., Gosselin et al., 2020; Ross et al., 2020; Zal et al., 2020). Here, upward fluid migration due to fault valving (Sibson, 1992; W. Zhu et al., 2020) may have aided not only the nucleation but also the rupture and tsunami complexity of the 2024 Noto events. The permeability of

294 the Noto fault system could have been low during its late interseismic period, allowing
 295 high pore-fluid pressure to effectively weaken the fault (Madden et al., 2022; Rice, 1992).

296 Well recorded moderate and large earthquakes have been shown to rupture complex
 297 fault networks in a variety of tectonic settings, involving subevents with distinct fault
 298 geometries (Hamling et al., 2017; Jia et al., 2023; Taufiqurrahman et al., 2023; Xu et al.,
 299 2023). We find that the Noto earthquake included six subevents rupturing multiple fault
 300 segments with different configurations: while the first five subevents likely break faults
 301 dipping towards the southeast direction, subevent E6 occurs on a northwest-dipping fault.
 302 Such complexity may reflect the complicated regional tectonic setting transitioning be-
 303 tween right-lateral strike slip faults and thrust faults in proximity to the Toyama Trough
 304 (Ishiyama et al., 2017; Oike & Huzita, 1988). It also agrees with a two-segment finite-
 305 fault model, which include information on fault orientations (Okuwaki et al., 2024).

306 Our subevent model demonstrates that resolving the moment release and associ-
 307 ated fault location and geometry is critical to inform tsunami rapid response efforts. Our
 308 tsunami simulation can explain the initial tsunami wave packets at most stations. How-
 309 ever, local discrepancies remain, including underestimating the observed tsunami heights
 310 at stations Fukaura (Fig. 3) and Toyama Bay (Fig. S11), which are likely due to (i) lim-
 311 ited resolution of bathymetry; and/or (ii) unmodeled effects of landslides. Bathymetry
 312 uncertainties are expected to have less impact on leading waves and their arrival times
 313 than on the trailing waves (Sepúlveda et al., 2020). Extensive landsliding has been re-
 314 ported shortly after the Noto Peninsula earthquake (Gomez, 2024; Matsushi, 2024), which
 315 may have locally affected the tsunami within Toyama Bay (Fujii & Satake, 2024; Koshimura
 316 et al., 2024; Masuda et al., 2024).

317 5 Conclusions

318 In this study, we unravel the complex rupture dynamics of the 2024 M_W 7.5 Noto
 319 Peninsula earthquake using a 6-subevent centroid moment tensor model that we obtain
 320 from teleseismic and strong motion Bayesian inversion. We observe two distinct rupture
 321 episodes: an initial, onshore rupture towards the southwest followed by a subsequent,
 322 partly offshore rupture towards the northeast, which re-nucleates at the earthquake's hypocen-
 323 ter after a 20-second delay and causes significant seafloor uplift releasing 40% of the to-
 324 tal seismic moment. Using the complex subevent model to simulate the resultant coastal
 325 tsunami yields large tsunami waves of up to 2.71 m in the source region. Our simula-
 326 tion accurately captures tsunami first arrival timing and overall wave amplitudes. Upon
 327 comparison with alternative source models, our findings imply the necessity of using ac-
 328 curate earthquake models that incorporate realistic fault geometries for rapid tsunami
 329 modeling and early warning.

330 Open Research

331 The original tide gauge data are obtained from the Intergovernmental Oceanographic
 332 Commission (IOC; <http://www.ioc-sealevelmonitoring.org>; last access: 1 April 2024)
 333 and from the Geospatial Information Authority of Japan (GSI; https://www.gsi.go.jp/kanshi/tide_furnish.html; last access: 1 April 2024). GeoClaw has been used for
 334 tsunami modeling (Clawpack Development Team, 2023). Our teleseismic data are from
 335 EarthScope (formerly IRIS) DMC (Albuquerque Seismological Laboratory/USGS, 2014).
 336 Regional strong motion data comes from the NIED strong-motion seismograph networks
 337 K-net and KIK-net (Okada et al., 2004). Statsmodels (Seabold & Perktold, 2010) and
 338 ObsPy (Beyreuther et al., 2010; Krischer et al., 2015) were used for data processing,
 339 Matplotlib (Hunter, 2007) and the Generic Mapping Tools (Wessel, 2024) for plotting.
 340 The geodetic data are obtained from Nevada Geodetic Laboratory (<http://geodesy.unr.edu>,
 341 last access: 1 April 2024) and GEONET, which is operated by the GSI.

343 **Acknowledgments**

344 We thank Ryo Okuwaki, Ignacio Sepúlveda, Jorge Macías Sánchez, and Thomas Ulrich
 345 for fruitful discussions. We thank the IOC and the GSI for making the sea level record-
 346 ings at the tide gauges in the Sea of Japan freely available. The authors acknowledge
 347 funding from the National Science Foundation (grant nos. EAR-2225286, EAR-2121568,
 348 OAC-2139536, OAC-2311208, NSF EAR-2022441, EAR-2143413), from the European
 349 Union's Horizon 2020 research and innovation programme (TEAR ERC Starting; grant
 350 no. 852992) and Horizon Europe (ChEESE-2P, grant no. 101093038; DT-GEO, grant
 351 no. 101058129; and Geo-INQUIRE, grant no. 101058518), the National Aeronautics and
 352 Space Administration (grant no. 80NSSC20K0495) and the Green's Foundation at IGPP
 353 at SIO.

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Supporting Information for “Rapid earthquake-tsunami modeling: The multi-event, multi-segment complexity of the 2024 M_W 7.5 Noto Peninsula Earthquake governs tsunami generation”

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Overview

We show the slip distribution of the first (USGS-T) and second (USGS-T+G) finite-fault model of the U.S. Geological Survey (2024) in Fig. S1. Next, we show the comparison of observed and synthetic seismic waveforms as used for the multiple subevent centroid moment tensor (CMT) inversion (Figs. S2-S7). Additionally, we show the uncertainties in the Bayesian 6-subevent CMT solutions obtained with the Markov Chain Monte Carlo (MCMC) inversion (Figs. S8, S9). Fig. S10 shows the vertical surface deformation of the preferred multi-CMT solution (also shown in Fig. 2) compared against aftershocks and against the fault traces from Fujii and Satake (2024) and MLIT (2014). This figure also shows the vertical displacement of the minimum and maximum offshore uplift multi-CMT solutions. Figs. S11-S13 show additional comparisons of simulated tsunami waveforms using the the USGS models, the minimum and the maximum offshore uplift multi-CMT solutions.

Tables S1, S2, and S3 provide source parameters of the preferred multi-CMT solution, information about the six subevents mapped to the corresponding fault segments, and the nodal plane standard deviations of the multi-CMT subevent solutions, respectively. Tables S4 and S5 provide insight into the arrival times of the initial tsunami crest and maximum simulated tsunami amplitudes.

Movie S1 shows a 3D view of the fault segments from the preferred multi-CMT solution together with the 30-hour aftershocks. Movie S2 shows the simulated tsunami based on the preferred multi-CMT solution. The full animations are available online.

Data Set S1. We will upload our subsampled 2000 multi-CMT solutions and all data required to reproduce our tsunami simulations to an openly available Zenodo repository.

Movie S1. 3D view of the fault segments from the preferred multi-CMT solution together with the 30-hour aftershocks.

Movie S2. Tsunami propagating based on the preferred multi-CMT solution for the three hours of simulated time.

References

Fujii, Y., & Satake, K. (2024). Slip distribution of the 2024 Noto Peninsula earthquake (MJMA 7.6) estimated from tsunami waveforms and GNSS data. *Earth, Planets and Space*, 76(1), 44. doi: 10.1186/s40623-024-01991-z

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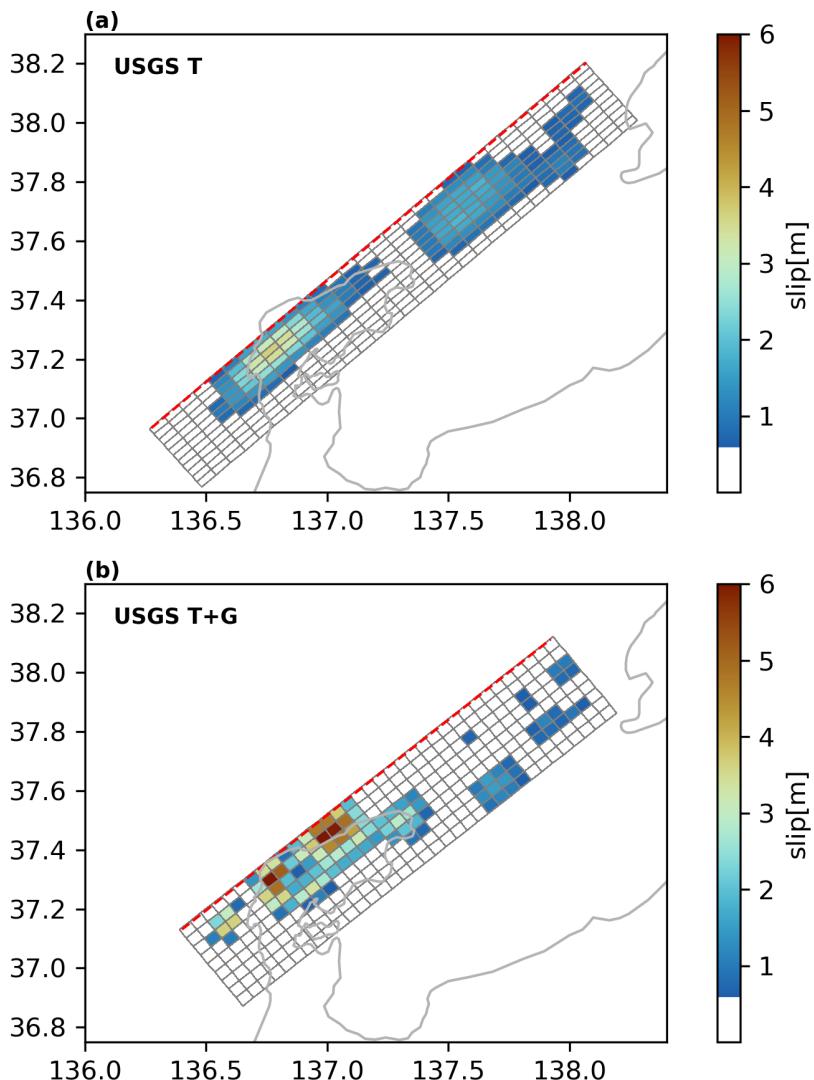


Figure S1. Slip distributions of (a) the first (USGS-T) and (b) the second (USGS-T+G) finite-fault model (U.S. Geological Survey, 2024).

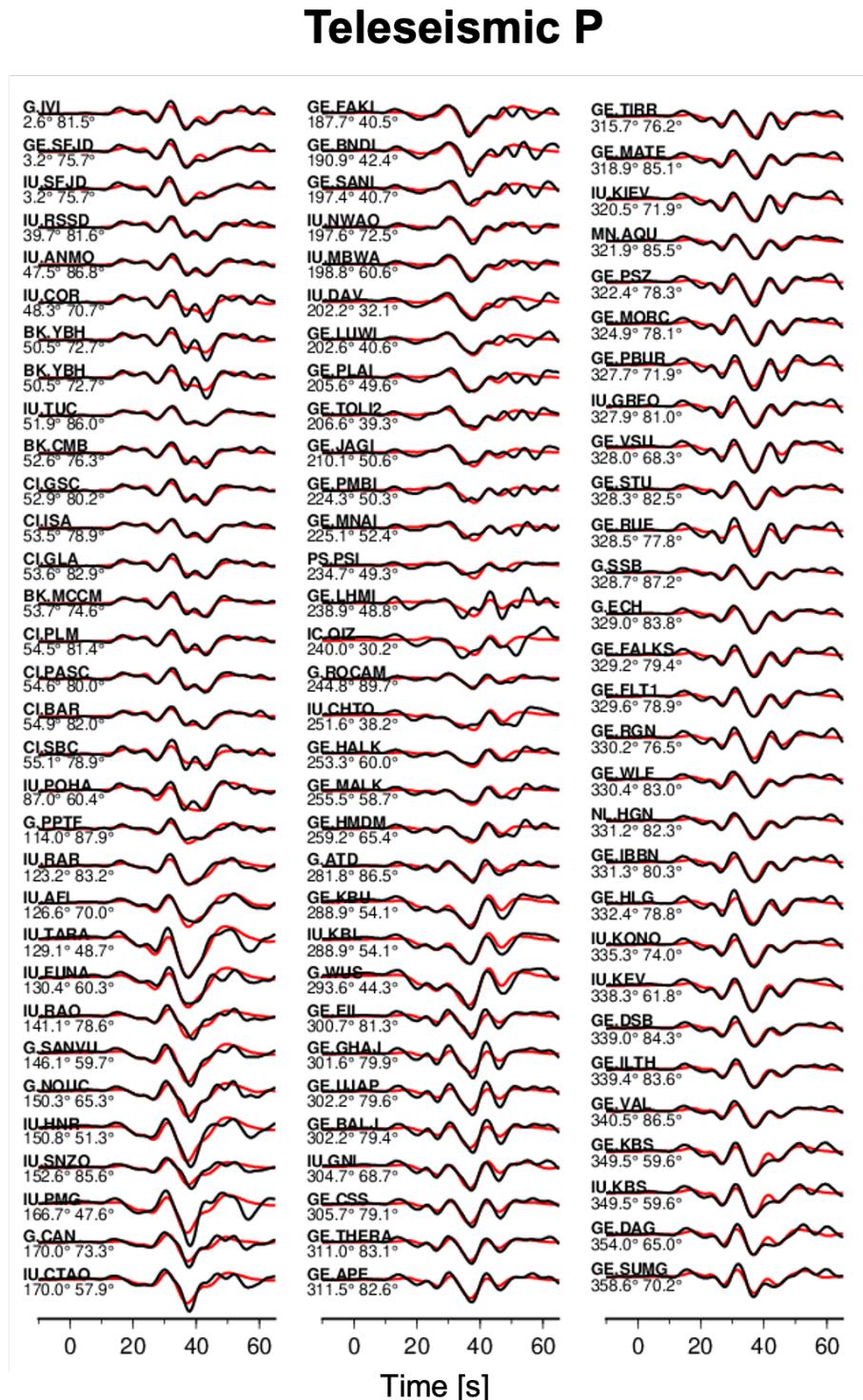


Figure S2. Observed (black) and synthetic (red) waveform fits for 93 teleseismic P waves in displacement. The numbers leading the traces are azimuths and distances.

Teleseismic P

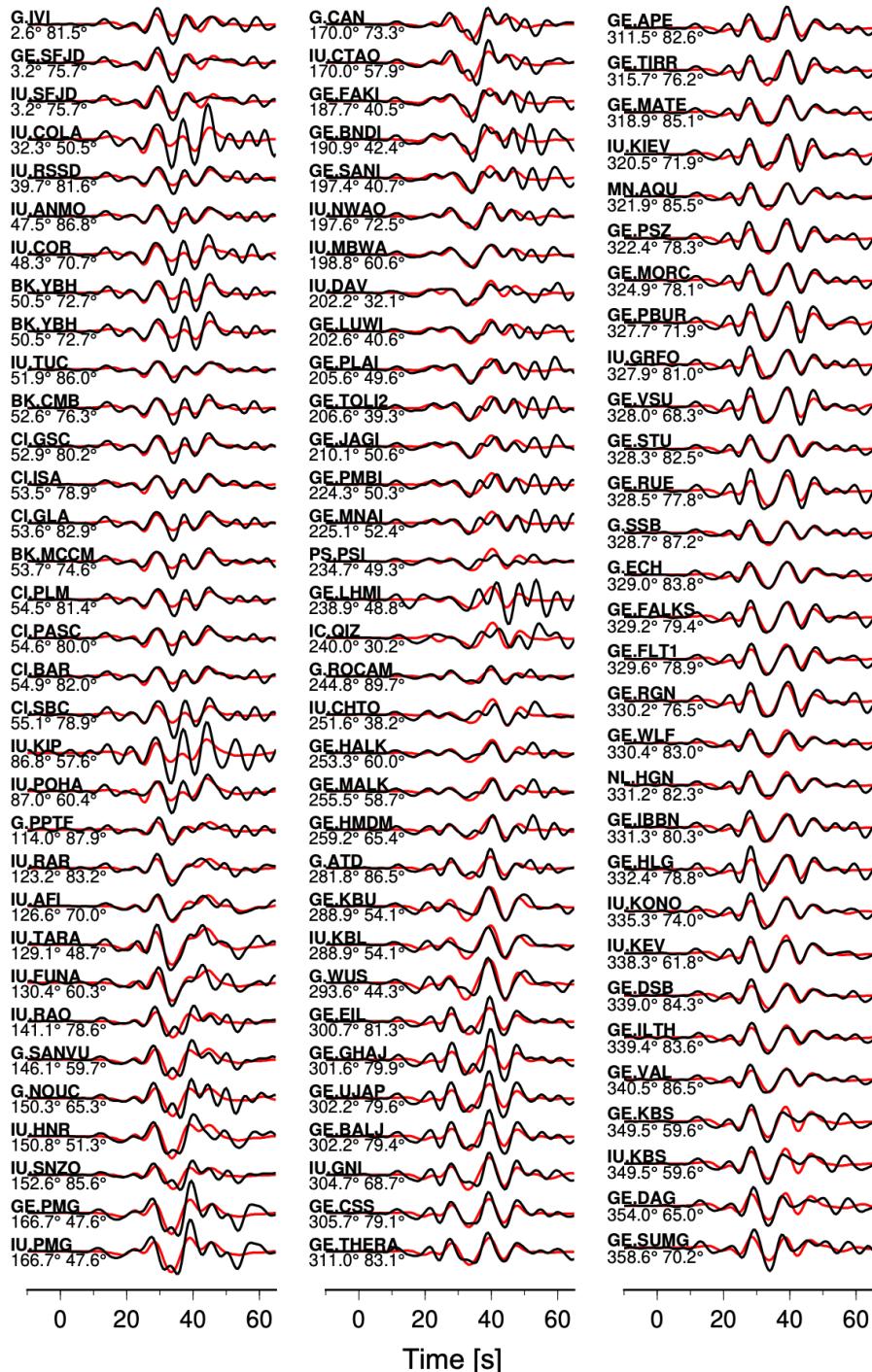


Figure S3. Observed (black) and synthetic (red) waveform fits for 93 teleseismic P waves in velocity, critical to fit the middle to high frequency content of the seismic radiation to far-fields. The numbers leading the traces are azimuths and distances.

April 13, 2024, 3:36pm

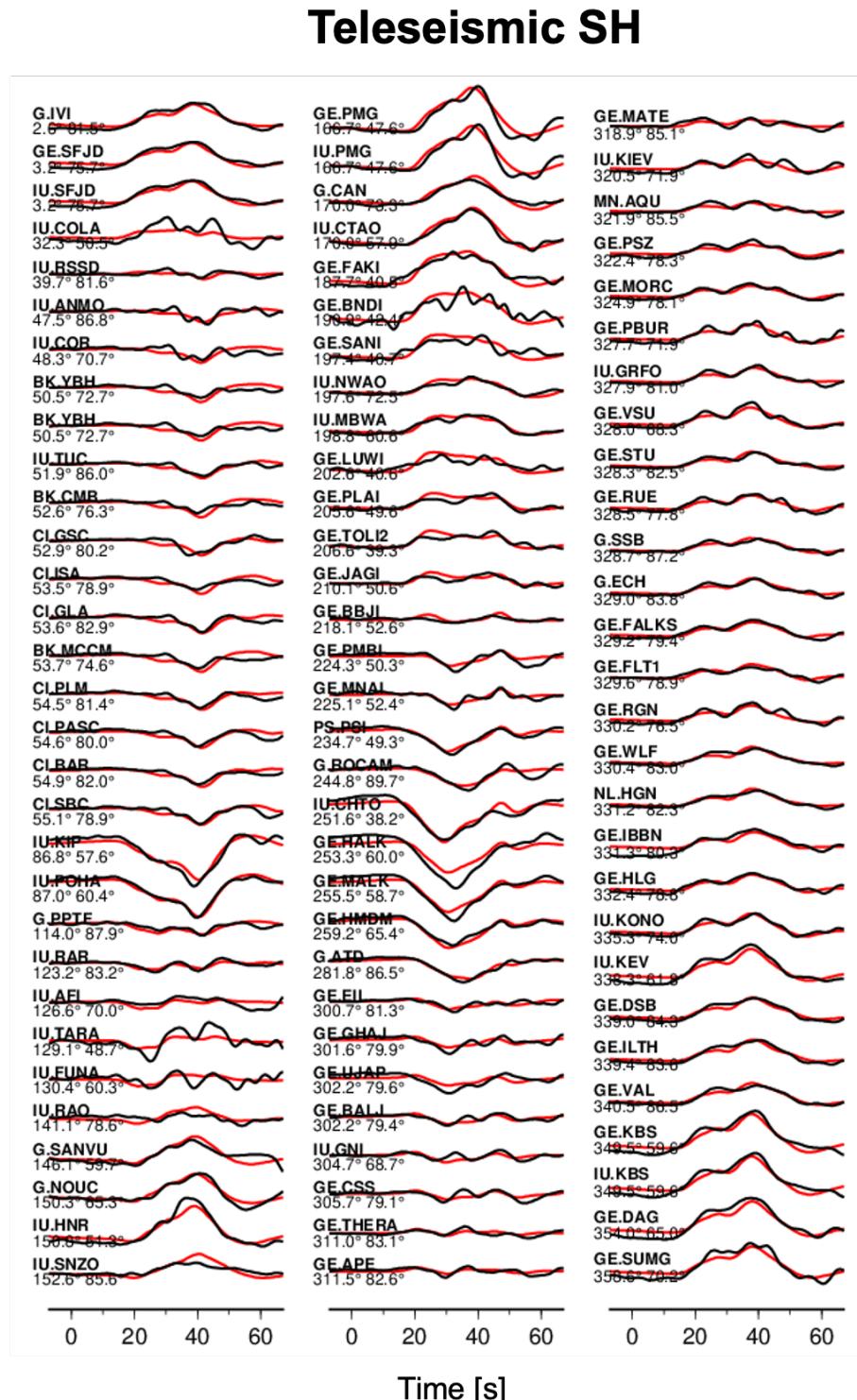


Figure S4. Observed (black) and synthetic (red) waveform fits for 93 teleseismic SH waves in displacement. The numbers leading the traces are azimuths and distances.

Strong motion waves

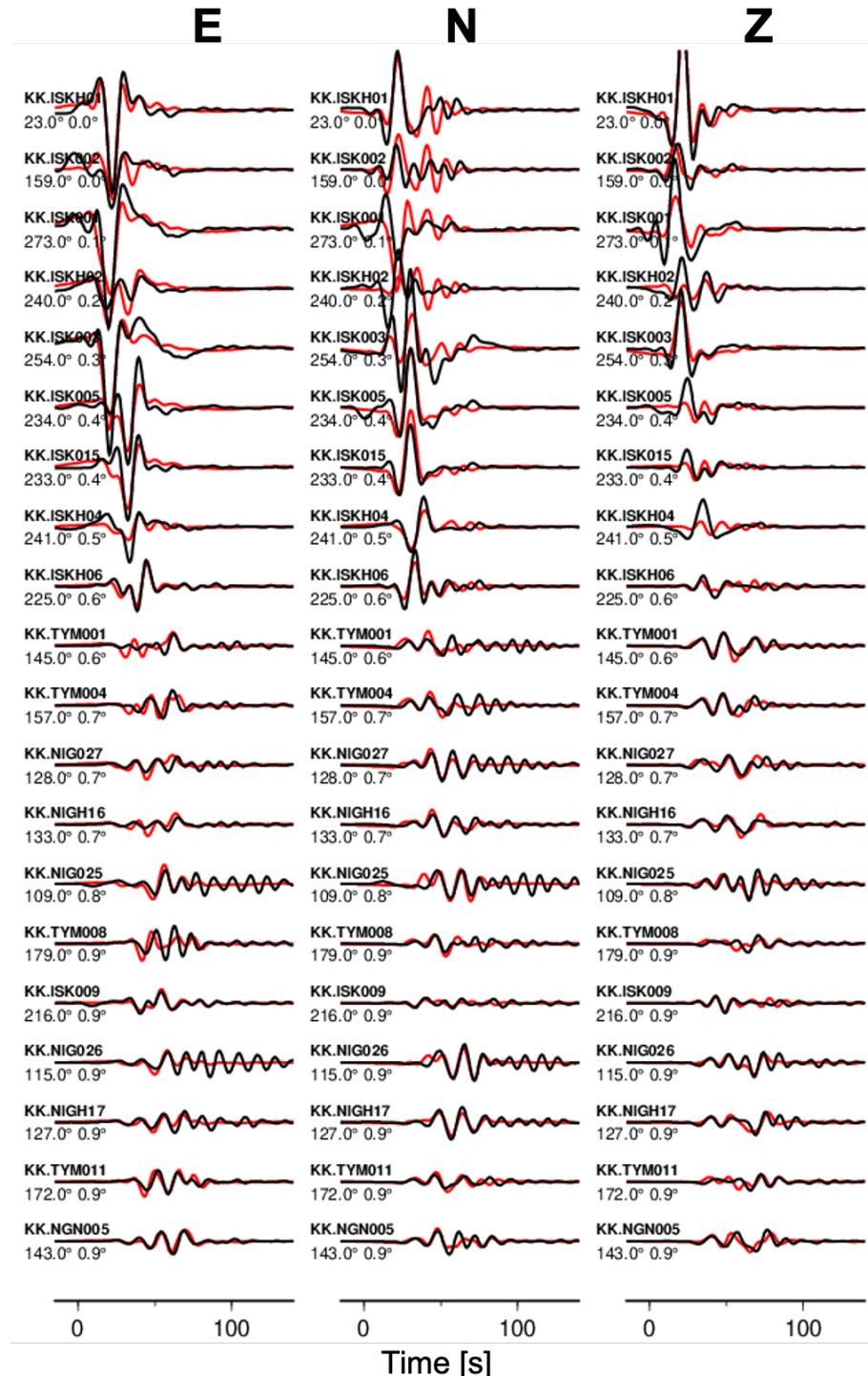


Figure S5. Observed (black) and synthetic (red) waveform fits for the first 20 (three-component) regional strong ground motion waves in displacement. The numbers leading the traces are azimuths and distances.

April 13, 2024, 3:36pm

Strong motion waves

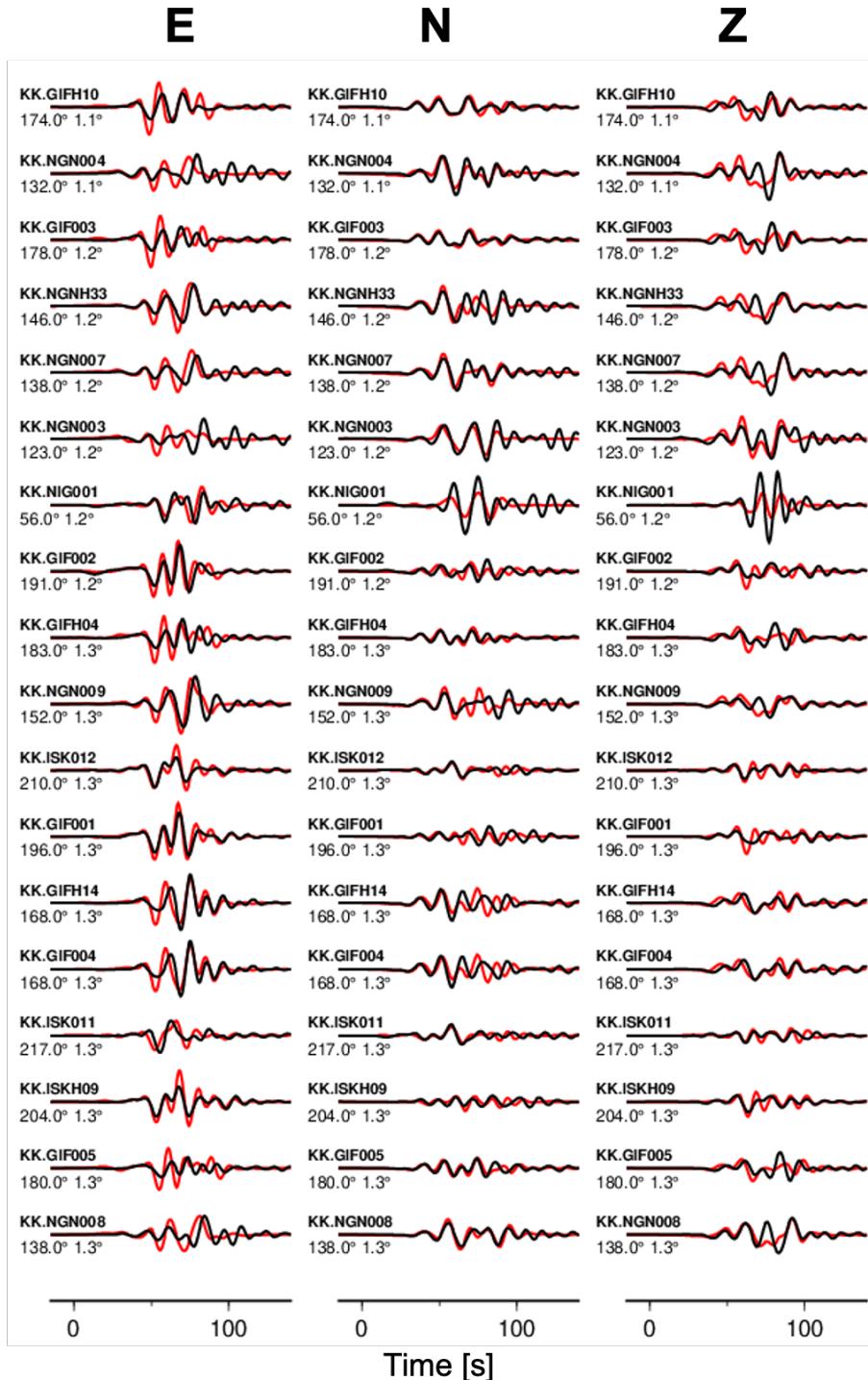


Figure S6. Observed (black) and synthetic (red) waveform fits for the second 20 (three-component) regional strong ground motion waves in displacement. The numbers leading the traces are azimuths and distances.

April 13, 2024, 3:36pm

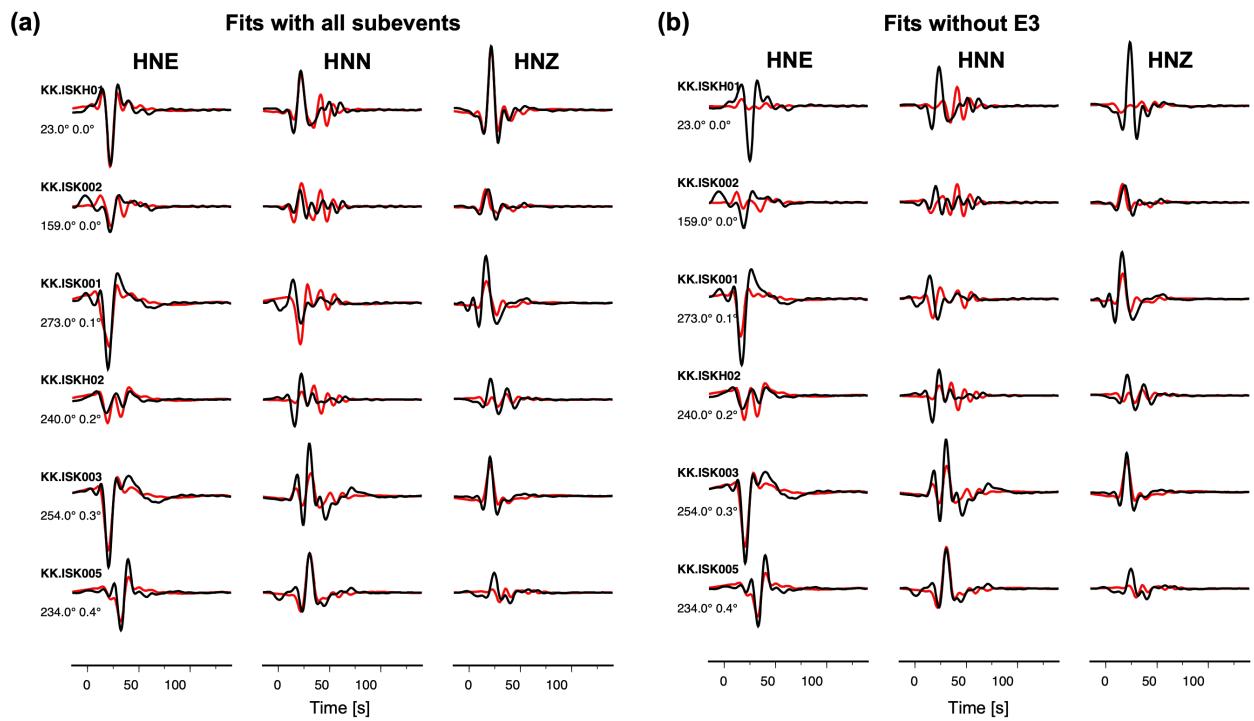


Figure S7. Comparison of strong ground motion waveforms with and without subevent E3.

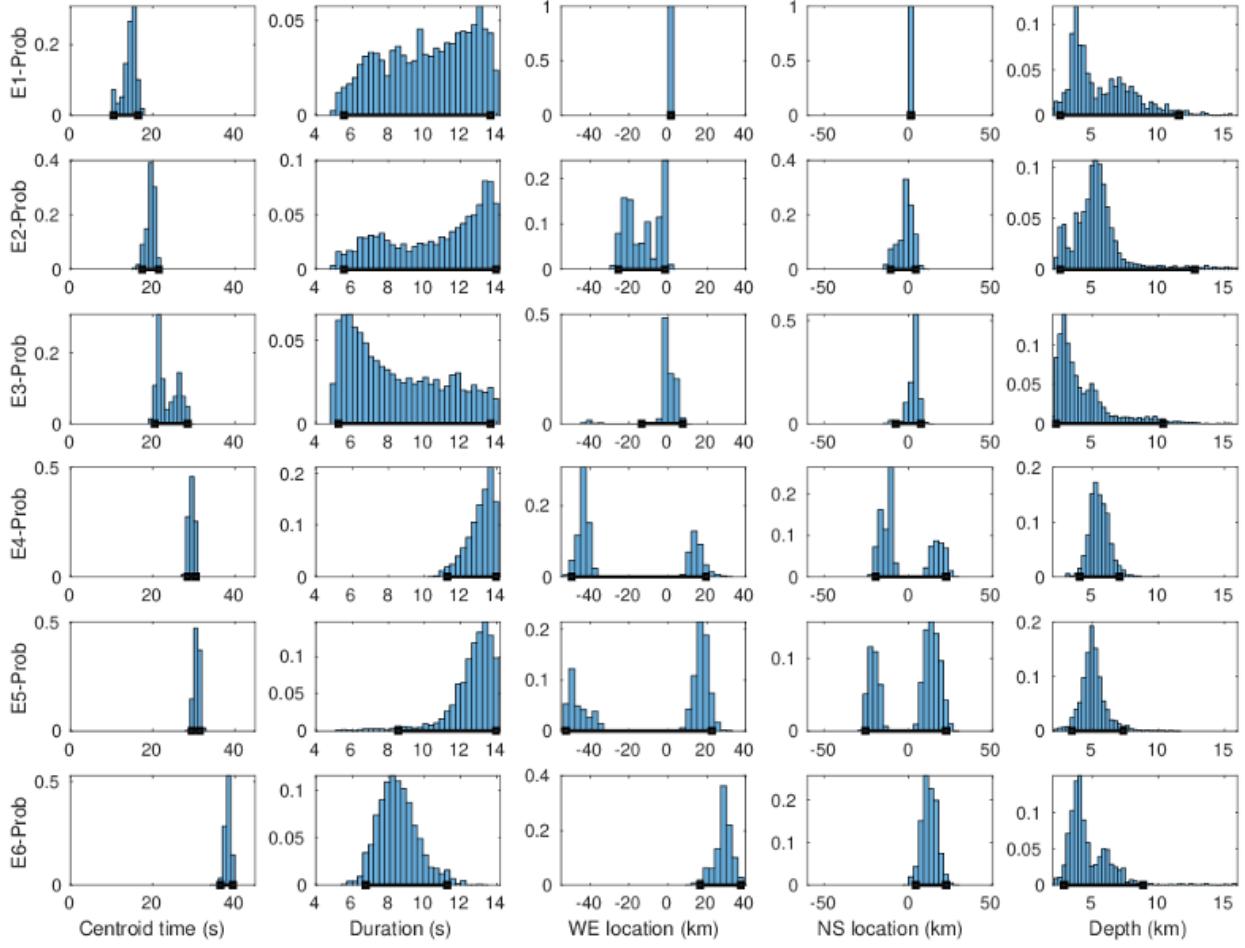


Figure S8. Centroid time, duration, West-East location, North-South location, and depth uncertainties for all six subevents (E1-E6) obtained from the multi-CMT inversion. Note that the location of the first subevent E1 is fixed at the JMA epicenter location, while its depth is allowed to vary. Black squares and lines indicate the error bars corresponding to the 95% confidential interval.

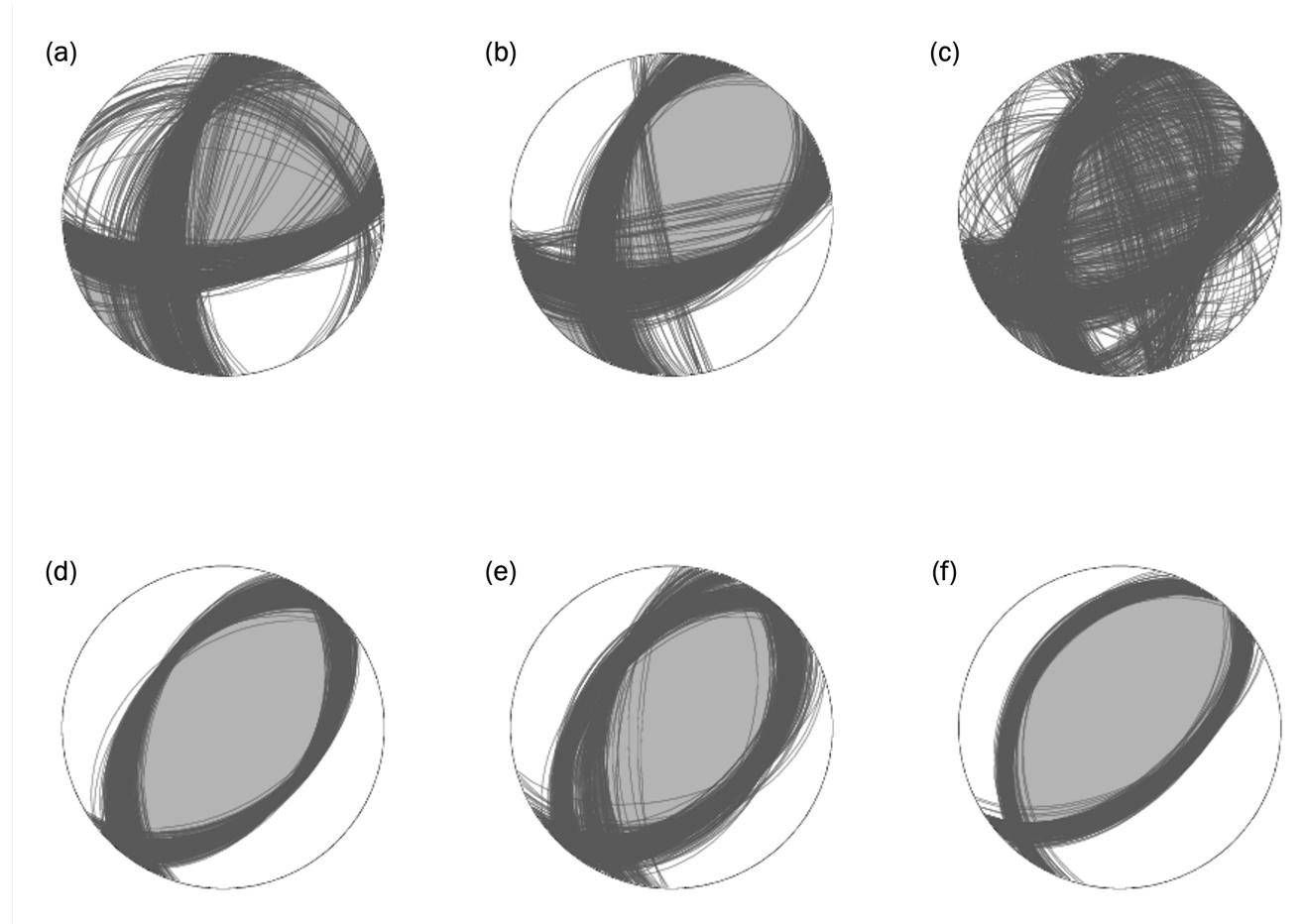


Figure S9. Fault and auxiliary plane uncertainties for all six subevents obtained from the multi-CMT inversion: (a) E1, (b) E2, (c), E3, (d) E4, (e) E5, and (f) E6.

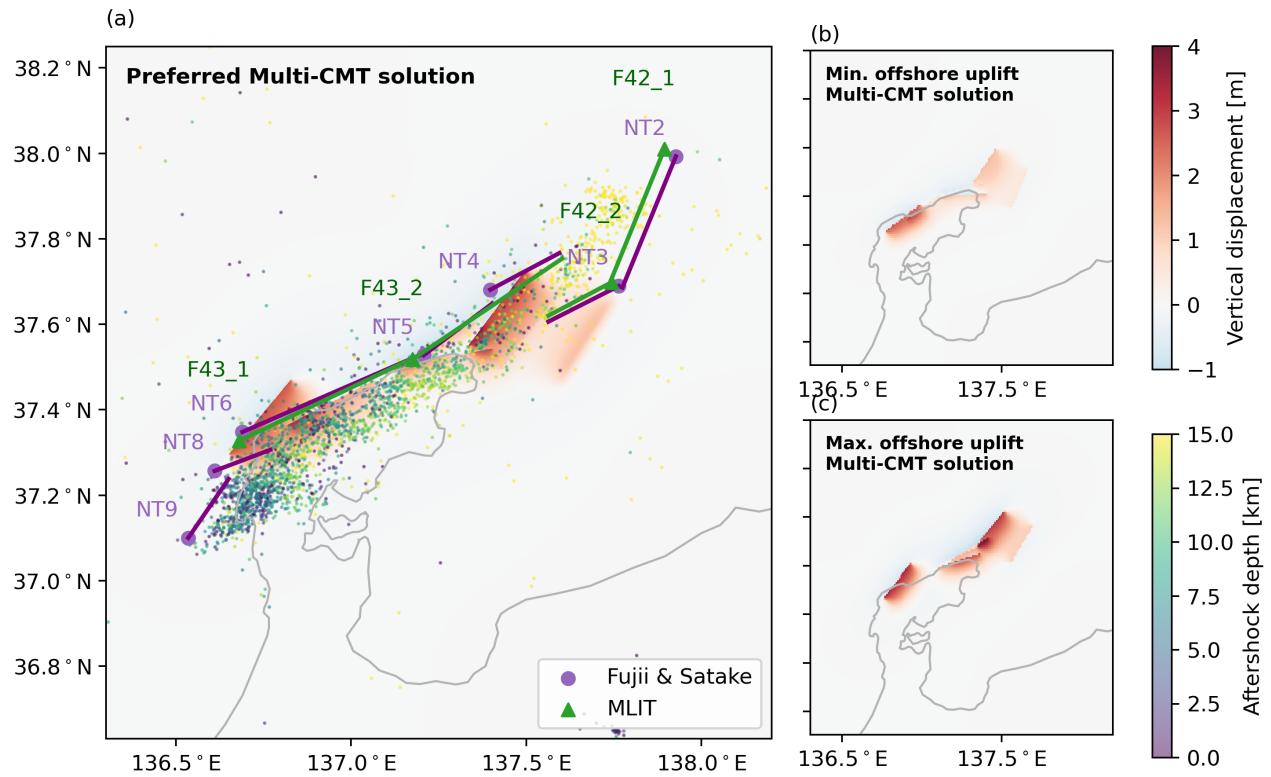


Figure S10. (a) Vertical surface deformation of the preferred multi-CMT solution (same as Fig. 2a) with the fault traces of Fujii and Satake (2024) and MLIT (2014) included. Panels (b) and (c) show the vertical displacement of the minimum and maximum offshore uplift multi-CMT solutions, respectively.

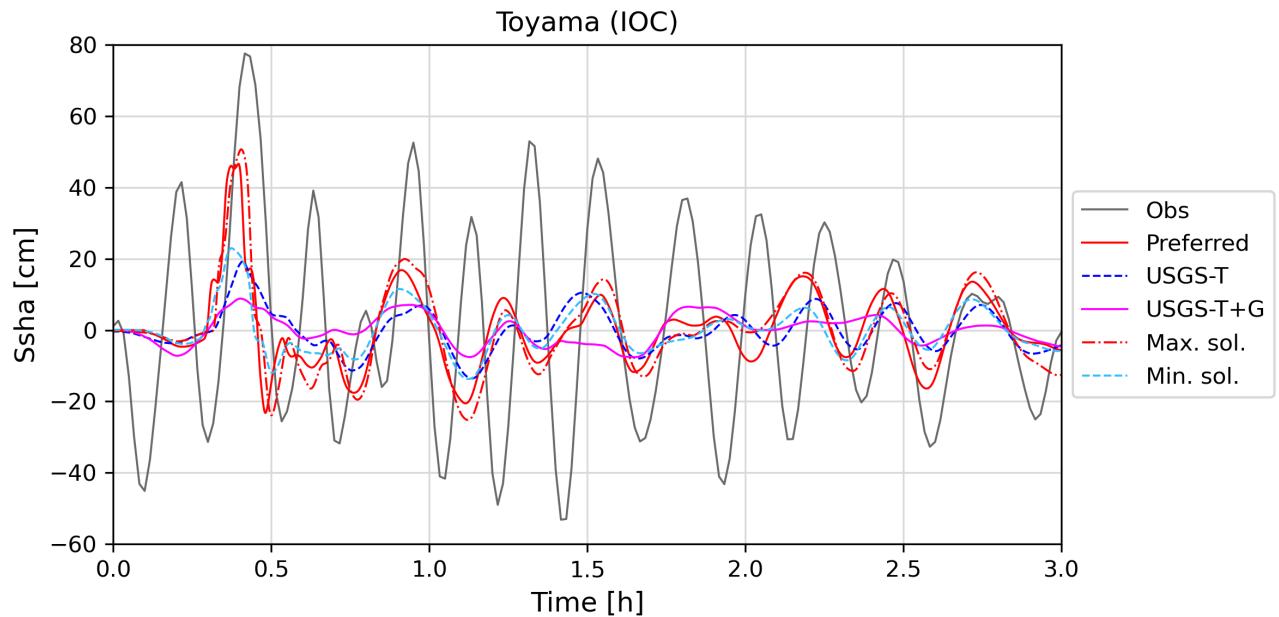


Figure S11. Comparison of observed and synthetic tsunami waveforms based on the preferred multi-CMT scenario at the tide gauge Toyama. The nearly immediate onset of the tsunami with negative polarity is likely related due to an additional contribution from a local landslide (Fujii & Satake, 2024; Koshimura et al., 2024; Masuda et al., 2024).

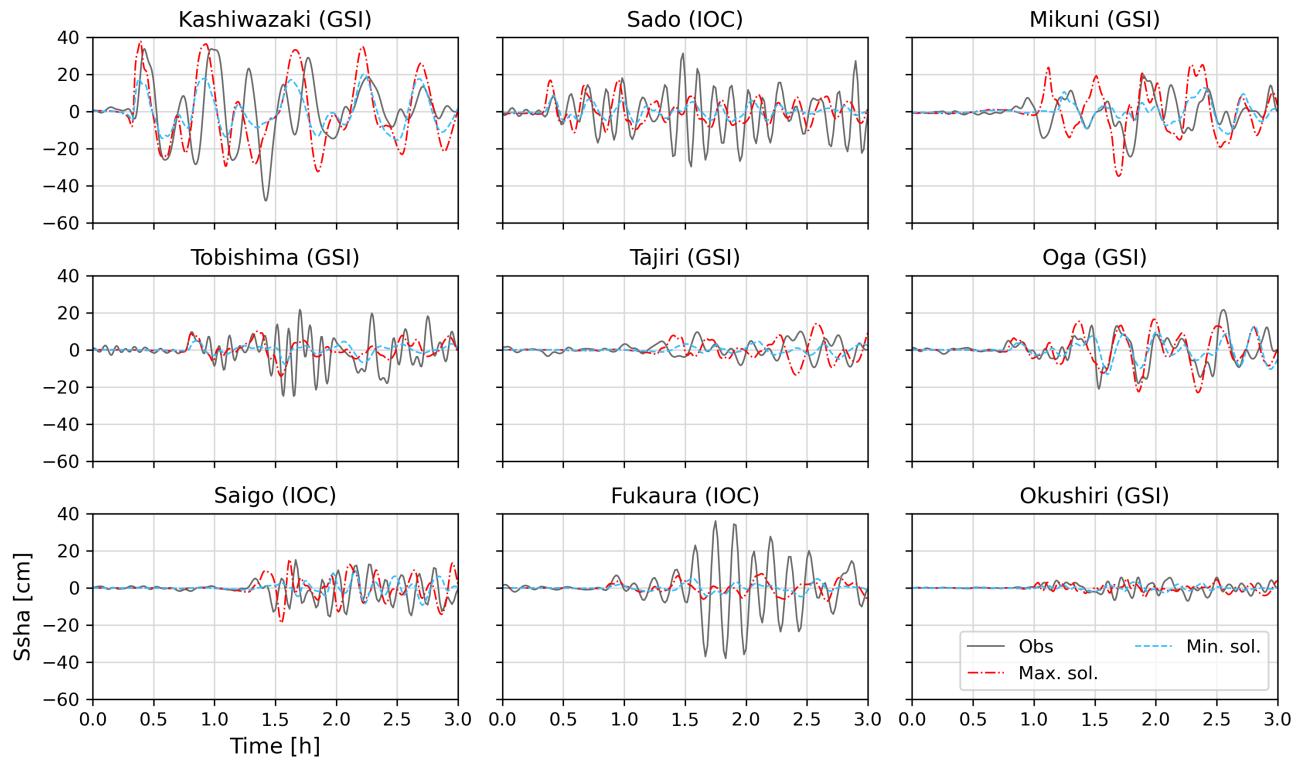


Figure S12. Comparison of observed and synthetic tsunami waveforms based on the minimum and maximum offshore uplift multi-CMT solutions at tide gauges facing the Sea of Japan, with stations sorted by their geodesic distance from the subevent E1.

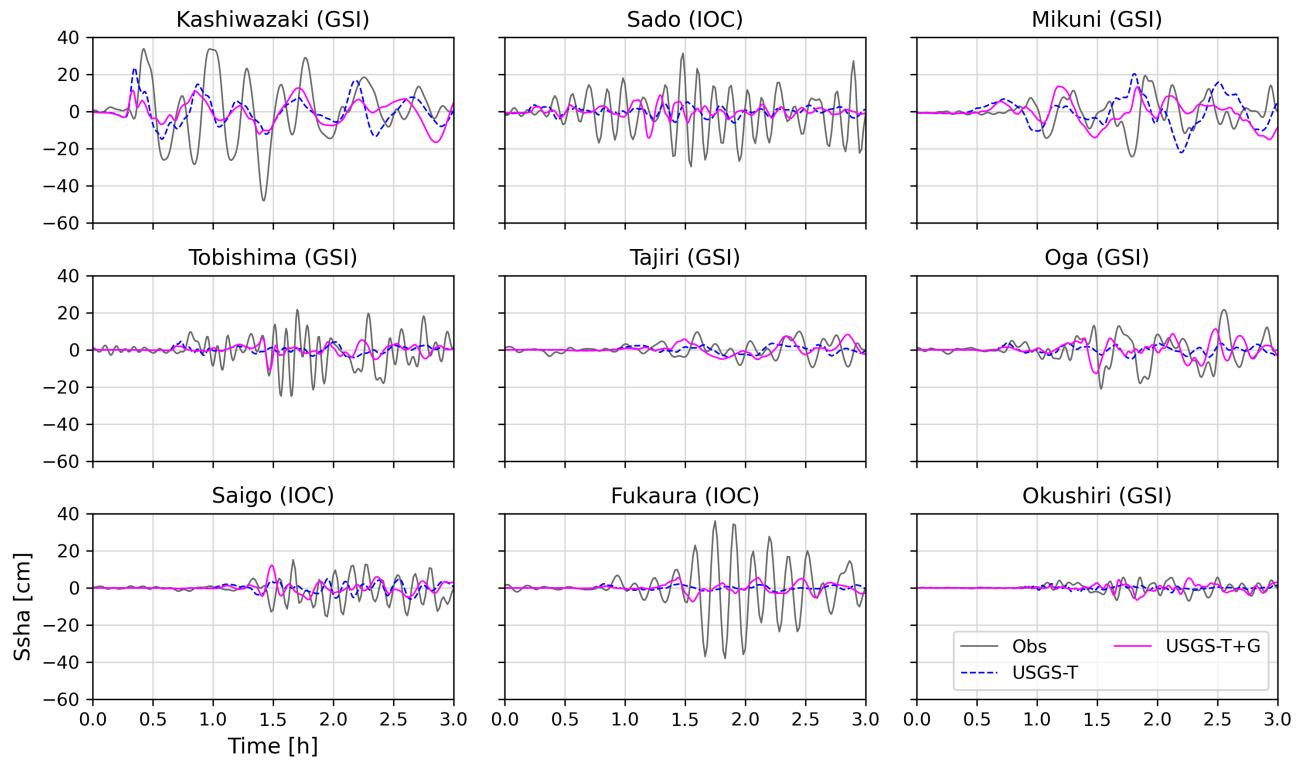


Figure S13. Comparison of observed and synthetic tsunami waveforms based on the USGS-T and USGS-T+G models at tide gauges facing the Sea of Japan, with stations sorted by their geodesic distance from the subevent E1.

Table S1. Preferred multi-CMT solution with its six subevents, E1 (top) to E6 (bottom).

Latitude, longitude and depth describe the fault segment center coordinates. The units of the respective seismic moments are $1e27$ dyne-cm.

centroid time [s]	lon	lat	duration [s]	depth [km]	Mw	Mxx	Mxy	Mxz	Myy	Myz	Mzz
15.447	137.27	37.495	11.058	4.668	6.9	-0.0795	0.1924	0.0806	-0.0550	0.1135	0.1345
19.883	136.966	37.3833	13.965	5.355	7.0	-0.136	0.1916	0.1016	-0.1512	0.0648	0.2871
21.552	137.25	37.5368	6.308	2.603	6.8	-0.0535	0.0920	0.066	-0.0929	0.0474	0.1465
29.612	136.798	37.3465	13.485	6.09	7.2	-0.1699	0.3699	0.0245	-0.4663	0.0952	0.6362
30.655	137.463	37.6099	13.229	5.242	7.2	-0.2705	0.3351	-0.0073	-0.4127	-0.0043	0.6833
38.496	137.614	37.5901	8.947	6.33	6.9	-0.1428	0.1629	0.1146	-0.1555	-0.0409	0.2983

Table S2. Preferred multi-CMT solution, with the six subevents mapped to the corresponding subfaults, E1 (top) to E6 (bottom). Latitude, longitude and depth describe the fault segment center coordinates.

lon	lat	depth [km]	strike	dip	rake	length [km] (along-strike)	width [km] (along-dip)	slip [m]
137.270	37.495	4.67	81.6	56.8	145.8	25.0	11.2	2.75
136.966	37.383	5.36	64.6	51.3	119.0	25.0	13.7	2.80
137.250	37.537	2.60	62.8	51.5	124.2	25.0	6.7	3.20
136.798	37.346	6.09	41.5	42.7	100.9	25.0	18.0	4.58
137.463	37.610	5.24	38.3	45.0	89.0	25.0	14.8	5.27
137.614	37.590	6.33	214.1	36.0	74.0	25.0	15.3	2.47

Table S3. Nodal plane standard deviations of the multi-CMT subevent solutions, E1 (top) to E6 (bottom)

dip	strike	rake
9.23	46.37	67.68
8.57	28.70	34.37
14.74	88.96	101.11
1.26	4.90	6.88
2.95	8.78	13.24
2.27	6.69	5.99

Table S4. Observed and simulated arrival times of the initial tsunami crest at the tide gauge locations. The arrival time is given in minutes after origin.

Tide gauges	Observed [min]	Preferred model [min]
Kashiwazaki	25.5	24
Sado	25	24
Mikuni	56.5	55.5
Tobishima	48.5	50
Tajiri	78.5	77
Oga	47.5	51
Saigo	80	77
Fukaura	56	58
Okushiri	65	62.5

Table S5. Maximum wave amplitudes for each of the tsunami simulations as shown in Fig. 4.

	Preferred multi-CMT solution	Min. offshore uplift multi-CMT	Max. offshore uplift multi-CMT	USGS-T	USGS-T+G
Max. wave amplitude [m]	2.71	2.38	3.36	1.25	2.38